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**Differences in motor unit behavior during isometric contractions in patients with
diabetic peripheral neuropathy at various disease severities**

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Abstract:

The aim of this study was to determine whether HD-sEMG is sensitive to detecting changes in motor unit behavior amongst healthy adults and type 2 diabetes mellitus (T2DM) patients presenting diabetic peripheral neuropathy (DPN) at different levels. Healthy control subjects (CON, n=8) and T2DM patients presenting no DPN symptoms (ABS, n=8), moderate DPN (MOD, n=18), and severe DPN (SEV, n=12) performed isometric ankle dorsiflexion at 30% maximum voluntary contraction while high-density surface EMG (HD-sEMG) was recorded from the tibialis anterior muscle. HD-sEMG signals were decomposed, providing estimates of discharge rate, motor unit conduction velocity (MUCV), and motor unit territory area (MUTA). As a result, the ABS group presented reduced MUCV compared to CON. The groups with diabetes presented significantly larger MUTA compared to the CON group ($p<0.01$), and the SEV group presented a significantly lower discharge rate compared to CON and ABS ($p<0.01$). In addition, the SEV group presented significantly higher CoV_{force} compared to CON and MOD. These results support the use of HD-SEMG as a method to detect peripheral and central changes related to DPN.

Keywords: Diabetic peripheral neuropathy; High-density surface electromyography; conduction velocity; Motor Unit; Motor unit territory area; Discharge rate.

Introduction

Diabetic peripheral neuropathy (DPN) is the most frequent diabetic neuropathy, being defined as a sensory-motor polyneuropathy, symmetric and distal. DPN represents 75% of neuropathy cases, causing sensory and motor deficits (Dyck et al., 2011, 1993; Pop-Busui et al., 2017; Selvarajah et al., 2019; Tesfaye et al., 2010). Motor deficits usually consist of force loss in distal lower limb muscles with a significant proportion of type I fibers (Oberbach et al., 2006), such as the tibialis anterior (TA) (Larsen et al., 2009; Oberbach et al., 2006; Regensteiner et al., 1998). Compromised TA function in DPN affects plantar pressure distribution, ultimately leading to bone deformities. Moreover, the association between bone deformities and dysfunctional plantar pressure distribution may lead to the development of plantar ulcers, considered the primary cause of lower-limb amputations in DPN patients (Fernando et al., 2013; Ferreira et al., 2017; Sacco and Sartor, 2016; Van Schie et al., 2004). Therefore, unraveling deficits in the muscle recruitment of DPN patients can assist in defining better strategies to avoid or at least delay major clinical interventions such as amputations.

Denervation of motor units (MU) is generally linked to force loss in degenerative diseases and aging (Allen et al., 2016; McNeil et al., 2005; Power et al., 2013). However, patients with DPN present a compensatory mechanism involving MU reinnervation to attenuate muscle weakness in the early stages of type II diabetes (Andreassen et al., 2006; Zochodne et al., 2008). The compensatory mechanism increases the MU territory, subsequently increasing muscle fiber density (Bril et al., 1996). Since the motor unit action potential (MUAP) propagation velocity is linearly related to the average diameter of its muscle fibers, the DPN mechanism of denervation and reinnervation may likely

influence motor unit conduction velocity (MUCV) (Blijham et al., 2006, 2004; Cruz-Martinez and Arpa, 1999).

Studies using intramuscular electromyography (iEMG) demonstrated chronic reinnervation in the TA muscle prior to sensory disturbances in type II Diabetes Mellitus (T2DM) patients. It has been shown that chronic reinnervation decreases MFCV while increasing abnormal single-fiber electromyography (*i.e.*, jitter) (Bril et al., 1996; Meijer et al., 2008; Shields, 1987). Moreover, T2DM and DPN patients present abnormalities in their action potentials and motor unit discharge rates in intrinsic muscles of the foot, hand, and the TA (Allen et al., 2013a, 2013b, 2015b). Furthermore, studies using murine models suggested that both MU denervation and sensory dysfunctions occur concomitantly at the initial stages of DPN (Ramji et al., 2007; Souayah et al., 2009). Muscle biopsies are often used to assess muscle properties and provide the extent of muscle degeneration (Gaster et al., 2001; Larsen et al., 2009; Oberbach et al., 2006). However, this technique is invasive, and its results do not inform the status of the muscular neural control properties, such as discharge rate or MFCV. Therefore, it is highly relevant to assess electrophysiological muscular properties of DPN patients non-invasively, as it can potentially become a method to quantify muscle degeneration and its neural control properties.

The relationship between MFCV and muscle fiber diameter has been previously established through iEMG recordings and muscle biopsies (Blijham et al., 2006). Therefore, EMG recordings can become an alternative method to access relevant information regarding muscle function and its neurophysiology, primarily if non-invasive methods are implemented. Currently, high-density surface electromyography (HD-

sEMG) provides a non-invasive method to identify large MU populations in isometric (Farina et al., 2008, 2001; Merletti et al., 2008; Negro et al., 2016) and dynamic contractions (Oliveira and Negro, 2021). HD-sEMG decomposition allows the estimation of motor unit discharge properties, MUCV with nearly negligible errors (<3%) (Farina et al., 2001), and motor unit territory area (MUTA) (Chandra et al., 2022; Gallina and Vieira, 2015; Kapelner et al., 2016). Therefore, HD-sEMG can be a relevant tool for assessing the properties of muscle recruitment.

The purpose of the present study was to determine whether HD-sEMG is a sensitive method to detect changes in motor unit discharge properties amongst healthy adults and patients at different stages of DPN. We hypothesized that DPN patients would present a reduction in the MUCV and discharge rate and that DPN patients would display an increased MUTA due to the reinnervation mechanisms leading to collateral branching of intact axons. Additionally, we hypothesized that the decrease in MUCV would be more pronounced in subjects with increased DPN severity. The compensatory mechanism that provides reinnervation is ceased at advanced stages of DPN, leading to muscle atrophy due to denervation or disuse (Andersen, 2012; Andersen et al., 1997; Meijer et al., 2008), ultimately reducing MUVc.

Materials and Methods

Participants

Thirty-eight adults with T2DM were recruited from the University Hospital of the Federal University of Santa Catarina. Diabetes patients were divided into four groups with different neuropathy severity: no DPN/absent group (n = 8); moderate DPN group (n = 18) and severe DPN group (n = 12). A simplified version of the NSS (Neuropathy Symptom Score) questionnaire was used to assess the main neuropathic symptoms. Moreover, a modified version of the mNDS test (modified Neuropathy Disability Score) was used to assess the main neuropathic signs. The classification of the severity of DPN was used as absent (symptoms and signs ≤ 2), moderate ($3 \leq \text{symptoms} \leq 7$, $3 \leq \text{signs} < 9$), and severe (symptoms ≥ 7 and signs ≥ 9) (Abbott et al., 2002; Moreira et al., 2005; Petropoulos et al., 2018; Young et al., 1993). Additionally, eight healthy matching adults (Control group) were recruited from the community.

Anthropometric and clinical data (diabetes mellitus duration, HbA1c test result, use of medications, and presence of complications) were obtained previously from the application of the study protocol. Inclusion criteria for patients: age >40 and < 70 years, T2DM was diagnosed based on the World Health Organization definition (World Health Organization, 2016). Exclusion criteria: minor/major limb amputation or other physical, neurological, musculoskeletal deficiencies, e.g., stroke, cerebral palsy, polio, arthritis, etc.; sight limiting diabetic retinopathy, severe nephropathy, chronic kidney disease stage 4-5, lower limb edema; active diabetic foot ulceration (Butugan et al., 2014). All study procedures followed the principles of the Declaration of Helsinki and were approved by the Human Research Ethics Committee of the Federal University of Santa Catarina

(Protocol Number: 2.390.994). Informed consent was obtained from all participants included in the study.

Experimental Protocol

Participants were instructed to be seated, with the hip and knee flexed at a 90° angle and the ankle in a neutral position at a 90° angle to the leg. The foot of the dominant leg was fixed on a custom dynamometer (Figure 1). After a familiarization session with the equipment and the experimental protocol, participants performed two maximum voluntary isometric contractions (MVIC) for the ankle dorsiflexion with a 5-s duration, interspaced by a 2-minute rest interval. Participants were verbally encouraged to produce the maximum effort throughout the task. The highest force achieved across the two MVICs was defined as the maximum force and used to define submaximal loads. Subsequently, participants performed one 20-s submaximal isometric ankle dorsiflexion at 30% MVIC. Real-time visual force feedback was provided to participants on a computer screen.

“INSERT FIGURE 1 HERE”

Data Acquisition

The HD-sEMG signals were recorded from the TA muscle using a 32-channel EMG system in a monopolar configuration. The sampling frequency was 2000 Hz, 8x gain, and digitized with a 24-bit A/D converter (Favretto et al., 2018). A 13x5 electrode array (ELSCH064NM2, OT Bioelettronica, Torino, Italy, 2 mm diameter, 8 mm inter-electrode distance) was used for the recordings. An 8x4 setup from the array provided the 32-channel EMG signals is shown in Figure 2. Before attaching the matrix, the skin was

cleaned and abraded. The EMG matrix was positioned according to published guidelines (Barbero et al., 2012) and fixed using adhesive tape. The electrodes' cavities formed due to the adhesive were filled with conductive paste (CC1, OT Bioelettronica, Turin, Italy). A reference electrode was fixed on the tuberosity of the tibia. The dynamometer measured the force using a strain-gauge load cell (Model SPL, traction/compression, 60-kg range, sensitivity 2 mV/V, AEPH of Brazil), digitized with a 24-bit A/D converter and a sampling frequency of 80 Hz. The reliability of this dynamometer has been described elsewhere (Andreis et al., 2019).

Data analysis

The acquired signals were analyzed using custom scripts in MATLAB (R2018b, MathWorks, MA, USA). The EMG signals were band-pass filtered (8th order Butterworth, 20-500 Hz) (Merletti, 1999), followed by a notch filter (2nd order, 60 Hz notch filter, and its subsequent five harmonics). The force signals were low pass filtered at 15 Hz using a fourth-order Butterworth filter. The HD-sEMG signals were decomposed into motor unit spike trains using an algorithm based on convolutive blind source separation (Negro et al., 2016). Only motor units with a decomposition accuracy (silhouette) greater than 0.9 were included in the analysis, followed by manual inspection (Boccia et al., 2019; Del Vecchio et al., 2020). The average discharge rate of the interspike interval (ISI) and the ISI coefficient of variation (CoV_{ISI} , ISI standard deviation divided by the ISI average) were calculated (Martinez-Valdes et al., 2015).

The decomposition algorithm identifies the discharge times of each motor unit, but not the waveform of the action potentials. To estimate the action potential waveform of each motor unit, the spike-triggered averaging technique was used. The motor unit action

potential (MUAPs) waveforms were extracted by averaging the monopolar HD-sEMG signals considering rectangular windows of 30 ms, using the discharge times as triggers. The average of the first 20 action potentials of each identified motor unit was used to estimate the motor unit action potential (Farina et al., 2002). From the identified monopolar action potentials, the differential double lead was computed, which is used to calculate the motor unit conduction velocity (MUCV). For the estimation of MUCV, a minimum of 3 and a maximum of 6 differential double leads were used. Double differential derivations are known to provide superior MUCV estimation, as they are less affected by the effects of end fibers and end plates (Farina et al., 2002). The selection criteria for the channels were clear propagation along with the columns of the motor unit action potential and a coefficient of correlation greater than 0.8 (Del Vecchio et al., 2018; Farina et al., 2002). The MUCV was estimated using the multichannel maximum likelihood algorithm, which calculates conduction velocity in MUs with a standard deviation of fewer than 0.1 ms⁻¹ (Farina et al., 2001).

“INSERT FIGURE 2 HERE”

The MUTAs were calculated using the RMS values of each action potential of the decomposed MUs were estimated using a simple differential configuration. The threshold value was selected based on the channel with the highest RMS amplitude within the grid. Then, only values exceeding 50% of the maximum RMS were used. The threshold of 50% value was selected according to the literature (Kapelner et al., 2016). Furthermore, it was observed that a threshold of 50% better demarcated the two-dimensional boundary of the MU territory, which was then segmented with an ellipse. The MUTA was expressed in terms of the number of channels in the electrode matrix (Chandra et al., 2022). In

addition, it was measured the coefficient of variation of the force (CoV_{force} , force standard deviation divided by the mean force) of the submaximal isometric ankle dorsiflexion at 30% MVIC.

Statistical Analysis

Statistical analyses were performed using R (R Code Team, 2018). Data are provided as mean \pm standard deviation. The normality of the dependent variables age, body mass index (BMI), MVIC, MVIC/body mass, glycated hemoglobin (HbA1c), DM duration, subcutaneous tissue thickness, MUCV, CoV_{ISL} , MUTA, discharge rate and CoV_{force} were assessed by the Kolmogorov-Smirnov test. Levene's test was used to test for homogeneity of variance across groups. The main effects of groups (Control x absent DPN x moderate DPN x severe DPN) were assessed for all variables (i.e., MUCV, CoV_{ISL} , MUTA, discharge rate, CoV_{force}) using a one-way ANOVA. The Tukey post-hoc test was used if necessary. For categorical data, Fisher's exact test was used. A significance level of 5% was adopted. The p -values, F -values, and the partial ETA squared (η^2_p) were reported.

Results

There were no significant between-groups differences in gender distribution ($p = 1$), age ($p = 0.26$), BMI ($p = 0.40$) and subcutaneous tissue thickness ($p = 0.64$, Table 1). Moreover, there were no significant between-group differences T2DM duration ($p = 0.84$) and HbA1c ($p = 0.49$), see Table 1.

There were main effects of the group for the mean MVIC (F -value = 7, $\eta^2_p = 0.33$, $p < 0.01$, Table 1) and MVIC/body mass (F -value = 5.3, $\eta^2_p = 0.28$, $p < 0.01$). The post-hoc analysis revealed that the severe group presented a 52% reduction in MVIC compared to

the control group ($p < 0.01$) and a 36% reduction compared to the moderate group ($p < 0.01$). Regarding MVIC/body mass, the post-hoc analysis revealed a reduction of approximately 41% for the severe group compared to the control group ($p < 0.01$).

“INSERT TABLE 1 HERE”

Illustrative data from MUCV and discharge rates from single individuals in every study group are displayed in Figure 3. The individual with severe DPN presents a reduced MUCV (Figure 3a) as well as reduced discharge rates across the reliably identified motor units compared to participants from the Control, Absent and Moderate groups (Figure 3b).

“INSERT FIGURE 3 HERE”

No between-group differences were found for CoV_{ISI} ($p = 0.8$). However, there was a main effect of the group for MUCV (F -value = 11.8, $\eta^2_p = 0.46$, $p < 0.01$, Figure 4(a)). The post-hoc analysis revealed that the MUCV from the severe group was lower compared to the control (average difference: -18%) and moderate group (-13%). Moreover, the absent group presented a 10% reduction in the MUCV compared to the control group. Regarding the MUTA, there was a main effect of the group (F -value = 4.2, $\eta^2_p = 0.23$, $p = 0.01$) (Figure 4(b)). Following the post-hoc analysis, the MUTA was larger for the absent (27 %), moderate (27 %), and severe groups (28 %) compared to the control group.

“INSERT FIGURE 4 HERE”

There was a main effect of the group for the mean discharge rate (F -value = 4.4, $\eta^2_p = 0.24$, $p < 0.01$) (Figure 5(a)). The post-hoc analysis exhibited that the discharge rate from the severe group was lower when compared to the absent (-19%) and the control groups (-18%). In addition, there was a main effect of the group for the CoV_{force} (F -value = 6.3, $\eta^2_p = 0.31$, $p < 0.01$) (Figure 5(b)). According to the post-hoc analysis, the CoV_{force} from the severe group was larger compared to the control (67%) and the moderate groups (37%).

“INSERT FIGURE 5 HERE”

Discussion

The purpose of this study was to determine whether HD-sEMG can be a suitable method to detect changes in motor unit recruitment properties in T2DM patients with DPN at different disease severities. The main findings of this study were that the group with absent DPN signs presented reduced MUCV and increased MUTA while maintaining discharge rate levels similar to controls. In addition, the moderate DPN group presented similar MUCV, discharge rate and CoV_{force} when compared to the controls, but exhibited greater MUTA. Furthermore, patients with severe DPN presented reduced MUCV and discharge rate, along with increased MUTA and CoV_{force} force compared to the control group. These results suggest that DPN may initially compromise only peripheral/morphological (i.e., motor unit conduction velocity) properties of muscle recruitment, leading to combined peripheral and central (i.e., motor unit discharge rate) deficits only at the later stage of the disease. Moreover, our study demonstrates that the decomposition of HD-sEMG signals is a relevant method for detecting changes in motor unit recruitment properties in T2DM patients presenting different DPN stages.

The reductions in force with the DPN progression in our study corroborate previous investigations (Andersen et al., 2004; Andreassen et al., 2006). Interestingly, it is possible to associate the changes in maximal force with muscle recruitment properties. Firstly, there is a reduction in force already in the absent DPN group, which is accompanied by a reduction in MUCV. MUCV reductions in T2DM patients presenting no signs of DPN may be explained by ongoing axonal losses and subsequent muscle fiber atrophy due to denervation and disuse (Andersen, 2012; Andersen et al., 1997; Meijer et al., 2008). From absent to moderate DPN, there is the maintenance of force levels, and the moderate group demonstrated similar MUCV. However, an increased MUTA was observed. Previous studies have described a compensatory reinnervation mechanism in DPN patients, in which denervated muscle fibers originally from low-threshold motor units are reinnervated by high-threshold motor units, resulting in an increased MUTA (Allen et al., 2015; Oberbach et al., 2006; Gaster et al., 2001 Andersen et al., 2009). This mechanism contributes to the maintenance of force levels and increases in the MUCV. Finally, the severe group generates even lower MVIC as a result of generalized muscle atrophy and neuronal losses (Meijer et al., 2008). Although the severe group also shows an increase in the MUTA, the lower MVC values may be related to the failure of the compensatory reinnervation mechanism and, consequently, generalized muscle atrophy (Allen et al., 2014; Andersen, 2012; Andersen et al., 1997; Meijer et al., 2008), which were illustrated in the present study by reduced motor unit discharge rate and motor unit conduction velocity.

Contrary to our initial hypothesis, MUCV did not gradually decrease as a function of DPN progression towards the severe stages. Instead, there were only specific significant reductions for the absent and the severe DPN groups. Interestingly, a reduction in MFCV

in the absent group has been previously reported (Butugan et al., 2014; Meijer et al., 2008; Suda et al., 2016). As previously mentioned, MUCV reductions in T2DM patients are related to axonal losses, muscle fiber denervation and disuse (Allen et al., 2014; Andersen, 2012; Andersen et al., 1997; Meijer et al., 2008), which underlines a peripheral adaptation in such patients. Moreover, T2DM is known to affect the composition of slow-twitch type I fibers (Gaster et al., 2001; Larsen et al., 2009; Oberbach et al., 2006). This compromise in muscle composition should be amplified in our study, as the TA muscle presents a greater proportion of type I fibers (Johnson et al., 1973). Therefore, the reduction in MUCV in the absence of DPN signs may be attributed to T2DM-related peripheral changes in muscle fiber content.

Interestingly, the MUCV from patients in the moderate group was generally the highest and statistically similar to the MUCV from patients in the control group (see Figure 4(a)). Indeed, Suda et al. (2016) reported an increase in MFCV in individuals presenting severe stages of DPN, which may be related to muscle fiber regeneration. Furthermore, it has been shown that T2DM may cause reductions in the proportion of lower diameter fibers with a corresponding increase in the proportion of higher diameter fibers as part of a compensatory reinnervation mechanism (Gaster et al., 2001; Larsen et al., 2009; Oberbach et al., 2006). Therefore, the increased MFCV in patients at the moderate DPN severity level may be related to the denervation of low-threshold motor units and the reinnervation of the available type of slow-twitch fibers (Suda et al., 2016) by high-threshold motor units. This result is highly relevant, as individuals with moderate DPN may be able to adapt to DPN-related deficits in motor performance through naturally occurring peripheral compensatory mechanisms. Future studies testing the effects of resistance/strength training on motor unit discharge properties in moderate DPN patients

can substantially contribute to minimizing motor deficits in DPN patients, potentially avoiding its progression towards severe DPN.

The motor unit discharge rate represents the descending command from the central nervous system to elicit muscle contractions (Heckman and Enoka, 2012), therefore considered a central component for motor unit function and control. As expected, the severe DPN group generated lower discharge rates during stable isometric contractions than the control and absent groups. This result corroborates previous studies investigating low-force isometric contractions for the TA (Allen et al., 2013b, 2014, 2015b) and vastus lateralis muscle (Watanabe et al., 2013) of patients with DPN. The decreased discharge rate in DPN patients has been attributed to changes in the contractile properties of muscle fibers (Allen et al., 2014; Watanabe et al., 2013), as well as to changes in Na⁺/K⁺ pump function resulting in conduction failure in patients with diabetic neuropathy (Krishnan et al., 2008). In addition, the results from our study demonstrated an increased MUTA in the TA muscle at all severity levels. This result suggests that a given force output may be reached by a smaller number of motor units recruited at lower firing rates (Allen et al., 2013b). The loss of motor units and the increase in MUTA may also explain the increase in force variability (CoV_{force}) (Figure 5(b)) for the severe group in relation to the control group (Figure 5(b)), as a smaller number of MUs may result in a less effective averaging process (Negro et al., 2009) and therefore poorer force control. This result corroborates the findings of previous research that reported greater variability at low levels of isometric dorsiflexion contractions in diabetic subjects with severe DPN (Suda et al., 2017) and

higher torque variability in individuals with type 2 DM without DPN (Senefeld et al., 2020).

Regarding asymptomatic DPN, increased variability in motor unit discharge rate has been found in patients with T2DM but asymptomatic to DPN, suggesting early neurophysiological DPN symptoms (Senefeld et al., 2020). However, our results regarding variability in inter-spike intervals did not reveal similar results. Nonetheless, our study demonstrated that non-invasive HD-sEMG was sensitive to underpin differences between healthy or patients with T2DM without DPN and those with considerable DPN progression, highlighting the potential of this technique to contribute to clinical research and practice.

The expected worsened performance for the severe DPN group has been confirmed by the reduced MVIC, MUCV, and discharge rate. In the severe stage of DPN, the denervation process is more effective than the previously mentioned reinnervation mechanism in place for moderate DPN, causing muscle atrophy and weakness due to the death of orphaned muscle fibers (Allen et al., 2016). Furthermore, previous studies have reported the loss of power, contractile quality, and muscle endurance in patients with DPN (Allen et al., 2015a, 2014; Hilton et al., 2008; Moore et al., 2016). In addition, there have been some suggestions on the association between T2DM and motor neuron dysfunctions that can ultimately lead to amyotrophic lateral sclerosis (Lekoubou et al., 2014; Logroscino, 2015; Sun et al., 2015). Motor neuron disorders may also influence

motor unit recruitment and compromise ideal motor unit firing during isometric contractions, partially explaining reductions in motor unit discharge rate.

A limitation of the study is the lack of HD-sEMG measurements of the ramp contractions performed to reach and leave the target force level (30% MVIC). Evaluating the ramp phases of the contraction would allow to investigate disease-related changes in rate coding and motor unit common input/synchronization across the T2DM severity groups. Future studies evaluating the rate coding at these different groups are necessary to elucidate the issue.

Conclusion

We conclude that individuals with DPN presented changes in the MUCV, MUTA, CoV_{force} and discharge rate. The T2DM patients with no signs of DPN already present neurophysiological signs of the disease presence, demonstrated by the reduced MUCV and increased MUTA for this patient group. Moreover, the moderate DPN group presented a similar MUCV and increased MUTA compared to patients in the control group, corroborating that reinnervation mechanisms may play a crucial role in maintaining motor function at this stage of the disease. Finally, the generalized detriment of motor unit recruitment properties in patients with severe DPN has been confirmed by the reduced discharge rate and MUCV, and increased MUTA and CoV_{force} . Therefore, we found evidence to support the use of HD-SEMG as a method to detect DPN-related changes in motor unit recruitment properties. However, more studies are needed to

improve our understanding of the neurophysiological adaptations caused by the increasing severity of DPN.

Declaration of Competing Interest

The authors declare that they have no known competing financial interests or personal relationships that could have appeared to influence the work reported in this paper.

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Captions to illustrations

Figure 1. Illustration of the experimental protocol to record the HD-sEMG signals.

Figure 2. Illustrative force and EMG-related signals from a type II diabetes patient without DPN (Absent group). In (a), the 64-channel matrix used for recordings, from which 32 channels (8x4 matrix, represented in the figure by the black rectangle) were used to record the electromyographic signals from the tibialis anterior muscle. In (b), the recordings consisted of force and 28 differentials signals acquired during isometric ankle dorsiflexion at 30% MVIC. In (c), action potentials propagation, discharge rate, and spike train of three motor units obtained after HD-sEMG signal decomposition.

Figure 3. (a) Motor unit conduction velocities (MUCV) defined from a sequence of surface EMG channels from single participants of the Control, Absent, Moderate, and Severe groups. (b) Discharge rates of decomposed motor units (MU) from single participants of the Control, Absent, Moderate and Severe groups.

Figure 4. Boxplot of motor unit conduction velocity (a) and motor unit territory area (b), for the different groups investigated. In all boxplots, center lines represent the median value, and the box limits illustrate the lower and upper quartiles (the 25th and 75th percentiles). The upper and lower whiskers extend to 1.5x the interquartile range, and data beyond the end of the whiskers are outliers, shown as dots.

Figure 5. Boxplot of motor unit discharge rates (a) and coefficient of variation of force (b), for the different groups investigated. In all boxplots, center lines represent the median value, and the box limits illustrate the lower and upper quartiles (the 25th and 75th percentiles). The upper and lower whiskers extend to 1.5x the interquartile range, and data beyond the end of the whiskers are outliers, shown as dots.

Table 1. Demographic and clinical data of the participants. Different letters represent statistically significant differences at $p < 0.05$.